

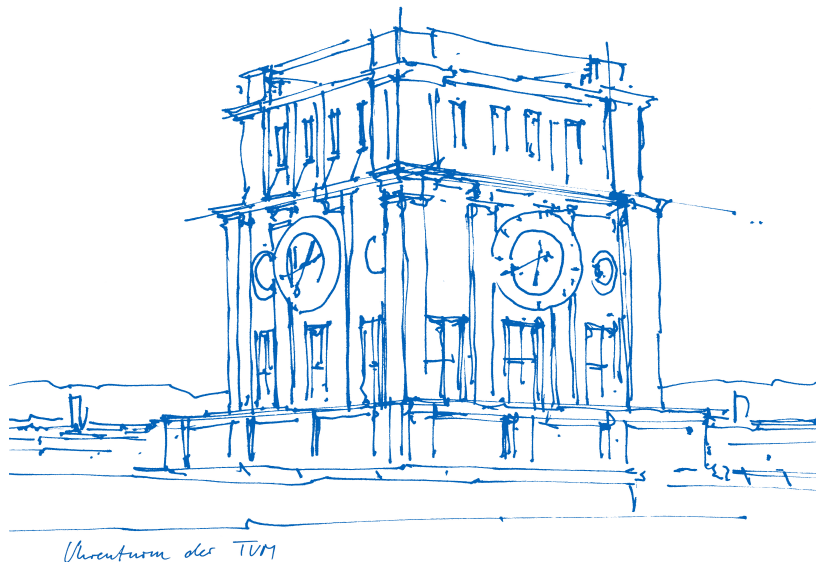
# An efficient MCMC algorithm for Subset Simulation

Seminars in Engineering Risk Analysis & Probabilistic Modelling

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15 June 2016



# Probability Puzzle



***Who is this?***

# Probability Puzzle – Answer

Typically associated with **Thomas Bayes** (1701?-1761)

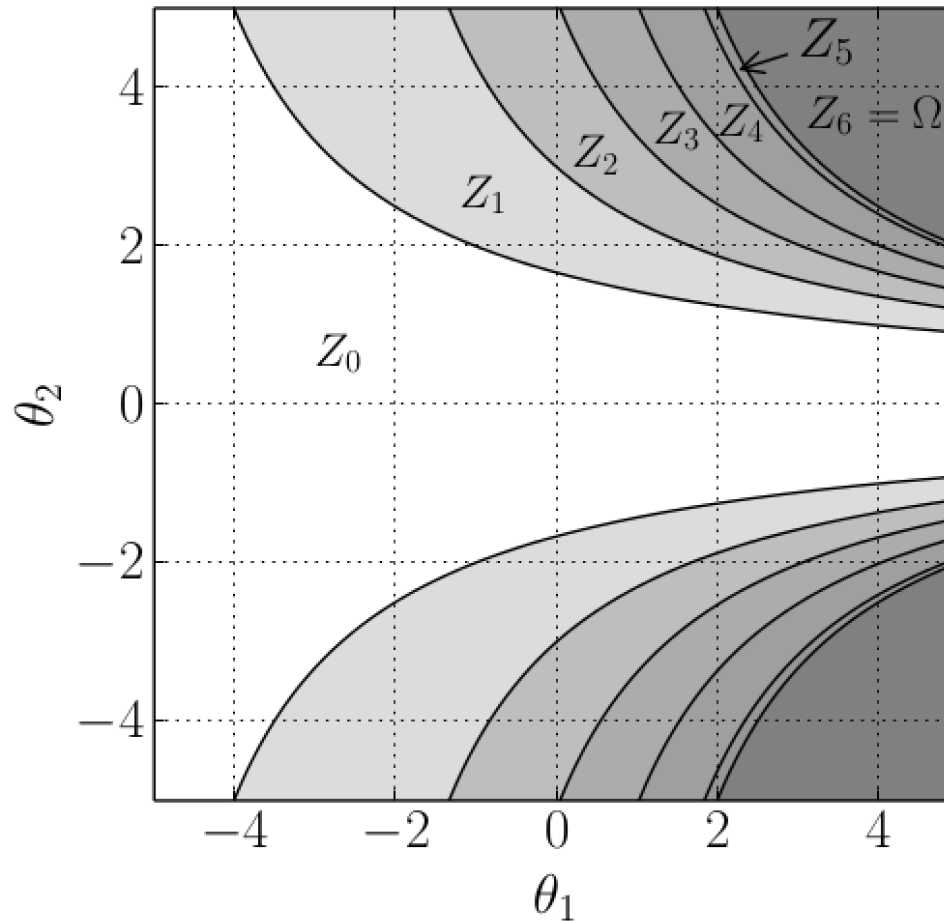
However, we do not know for sure!!!

David R. Bellhouse:

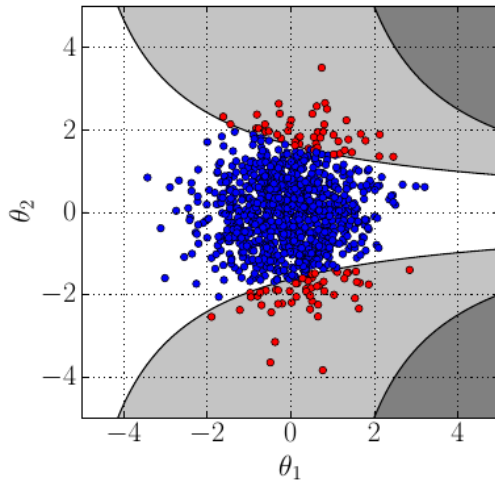
"The first thing to note in this picture is the apparent absence of a wig, or if a wig is present, it is definitely the wrong style for the period. [...] The second thing to note is that Bayes appears to be wearing a clerical gown like his father or a larger frock coat with a high collar [...] the gown is not in style for Bayes's generation and the frock coat with a large collar is definitely anachronistic. [...] For reference, I have used C. Willett Cunnington and P. Cunnington, Handbook of English Costume in the Eighteenth Century, pub. Faber & Faber, London, 1964."



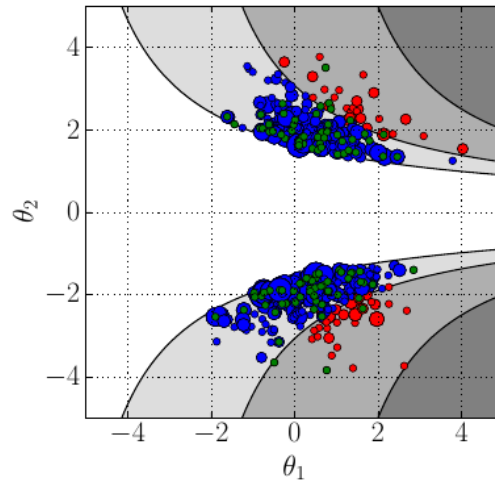
# Subset Simulation (SuS) – idea



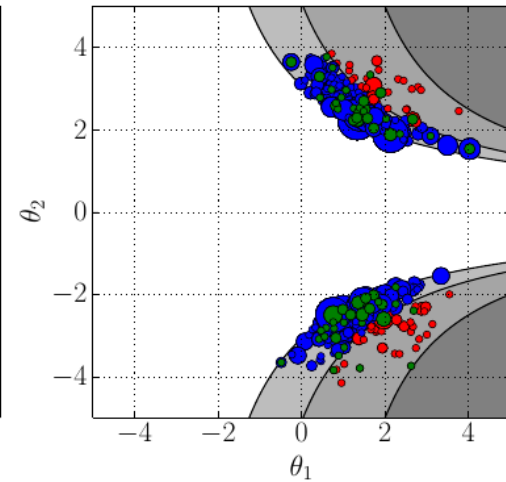
# Subset Simulation (SuS) – illustration



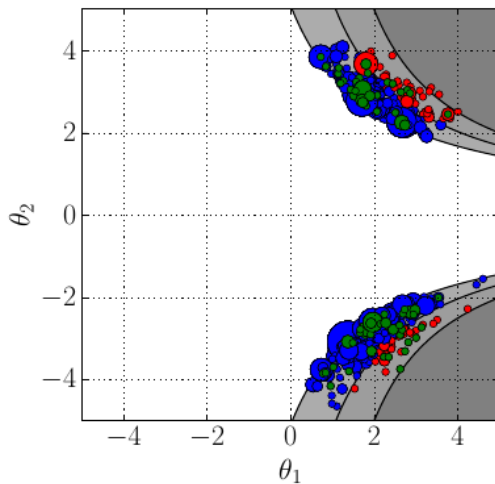
(a) level 0



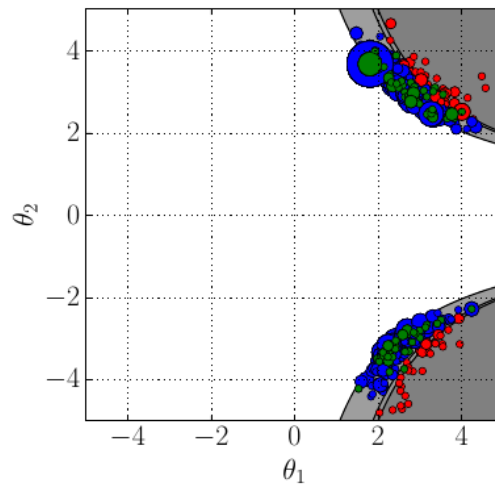
(b) level 1



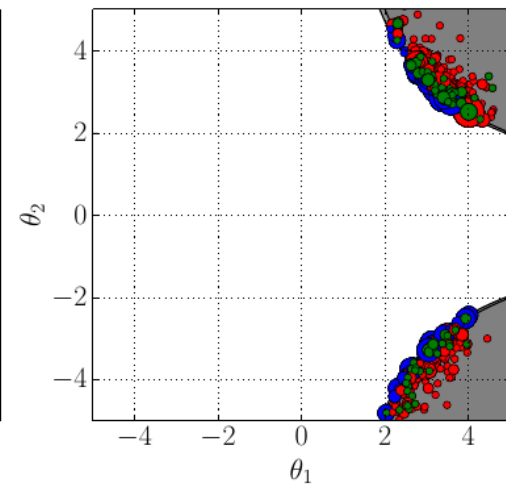
(c) level 2



(d) level 3



(e) level 4



(f) level 5

# MCMC in SuS: conditional sampling

[Papaioannou et al., 2015]

1. Let  $\mathbf{w}$  be a vector whose dimension is equal to the dimension of  $\mathbf{u}$ . For each component  $w_n$ ,  $n = 1, \dots, N$  of  $\mathbf{w}$  set  $w_n$  as a sample from distribution:

$$q(w_n | u_n^{(i,j,k-1)}) = \varphi \left( \frac{w_n - \rho_q \cdot u_n^{(i,j,k-1)}}{\sqrt{1 - \rho_q^2}} \right) \quad (3.38)$$

2. Evaluate  $g^*(\mathbf{u})$  and check if  $\mathbf{w} \in Z_i$ .

(a) If  $\mathbf{w} \in Z_i$ : set  $\mathbf{u}^{(i,j,k)} = \mathbf{w}$ ,

(b) otherwise set  $\mathbf{u}^{(i,j,k)} = \mathbf{u}^{(i,j,k-1)}$ .

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simple MCMC strategy

particularly efficient in high dimensions



MCMC in SuS: spread of proposal distribution

**What is the optimal spread?**



# MCMC in SuS: spread of proposal distribution

What is the **optimal** spread?

optimal spread: uncertainty in estimated  $p_f$  is **minimized**

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How do we **measure** uncertainty?

# MCMC in SuS: spread of proposal distribution






an  
What is ~~the~~ optimal spread?

optimal spread: uncertainty in estimated  $p_f$  is minimized



How do we measure uncertainty?

# MCMC in SuS – efficiency measures

-   $\text{eff}_{\text{MH},1} = \frac{\text{ESJD}}{\text{ESJD}_{\text{opt}}}$  expected squared jumping distance
-   $\text{eff}_{\text{MH},2} = \text{eff}_{\gamma, I_{G_1}}(\theta)$  indicator function (10%): eff. number of samples
-   $\text{eff}_{\text{MH},3} = \text{eff}_{\gamma, I_{G_2}}(\theta)$  indicator function (50%): eff. number of samples
-   $\text{eff}_{\text{MH},4} = \text{eff}_{\frac{1}{M} \sum_{i=1}^M \theta_i}$  variance of sample mean: eff. number of samples
-   $\text{eff}_{\text{MH},5} = \text{eff}_{\theta_1}$  variance of sample mean: eff. number of samples

# MCMC in SuS – Example 1

one-dimensional problem

truncated standard Normal target distribution

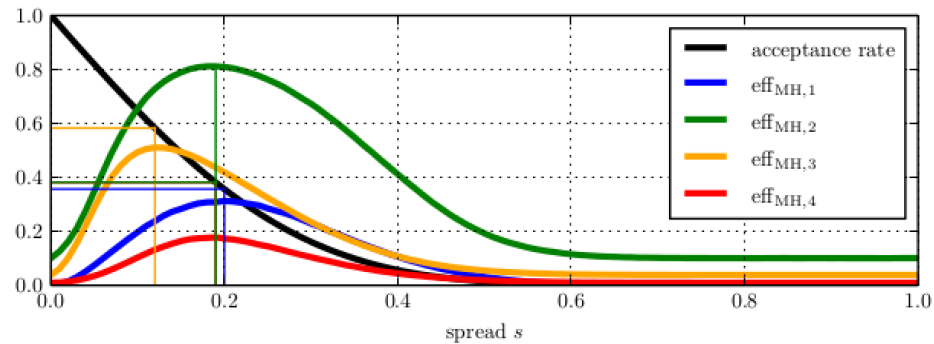
on support  $[9, \infty)$

rejection sampling acceptance rate:  $10^{-19}$

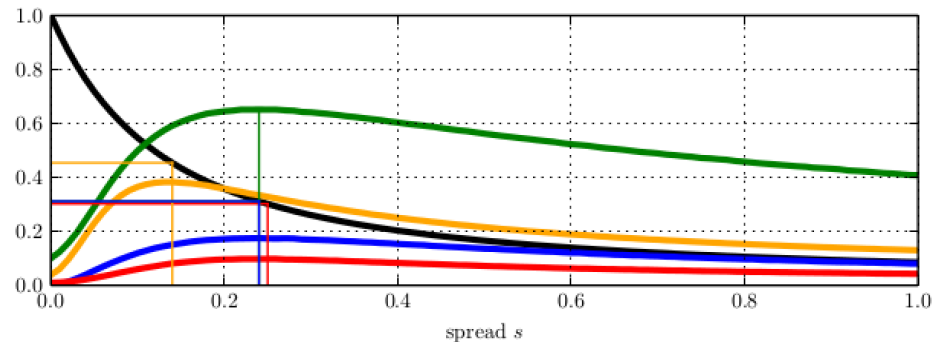
We investigate: optimal acceptance rates (and spreads) for different error measures.

# MCMC in SuS – Example 1

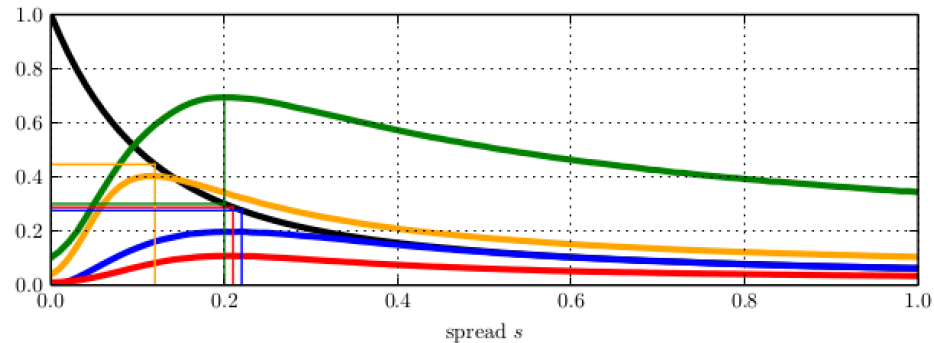
conditional sampling



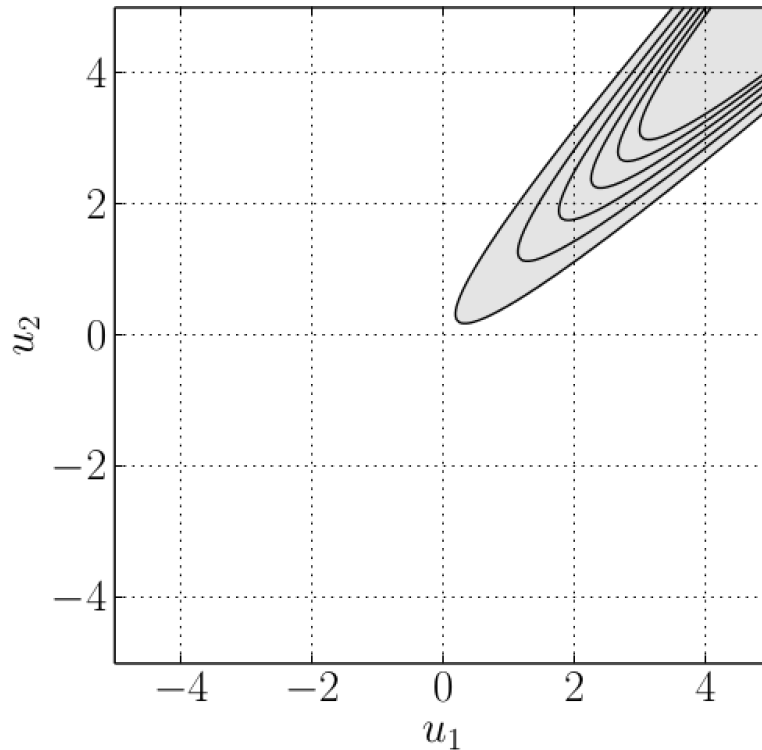
RWM (Normal)



RWM (uniform)



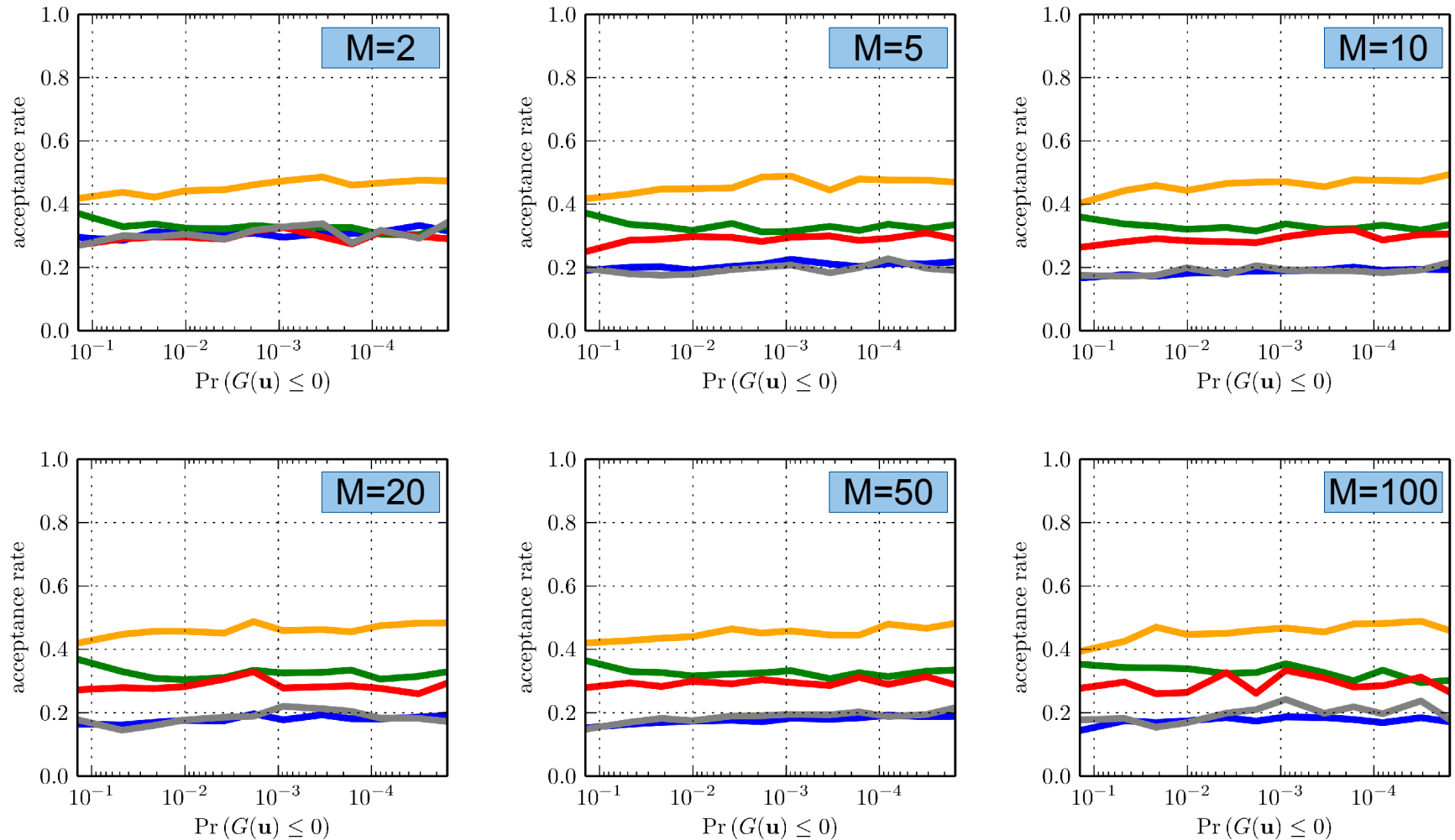
# MCMC in SuS – Example 2



$M = 10$

$$G(\mathbf{u}) = t - \frac{1}{M} \sum_{i=1}^M u_i + \frac{10}{4} (u_1 - u_2)^2$$

# MCMC in SuS – Example 2

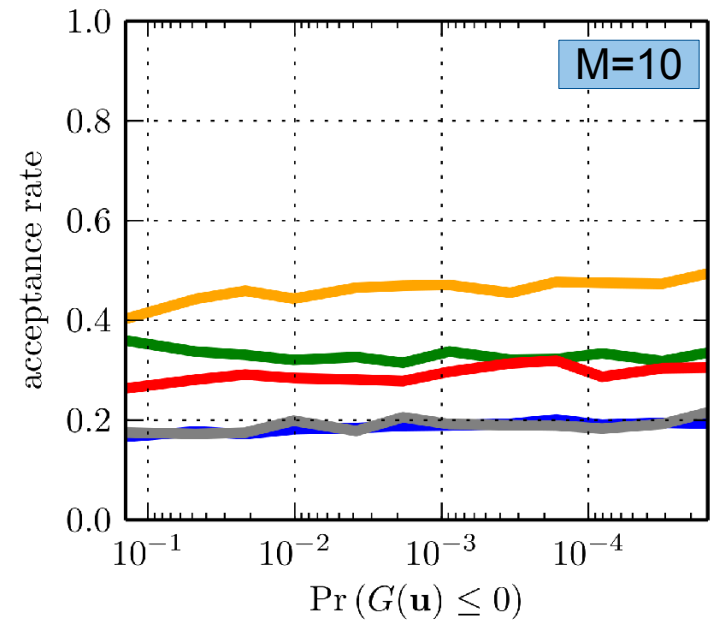




# MCMC in SuS – Example 2 (contd.)

optimal acceptance rate within SuS for Example 2:

target $p_f$	$N=10^3$	$N=10^4$
$10^{-5}$	0.29	0.24
$10^{-6}$	0.29	0.24
$10^{-20}$	0.24	0.21



# Summary

An acceptance rate of 0.44 is a robots choide.

However, finding an optimal acceptance rate can further improve the efficiency of SuS.

But ... the optimal acceptance rate depends on the investigated problem.

Moreover, it is not clear how to optimally measure MCMC efficiency in SuS.

Future work: Learn optimal acceptance rate through importance sampling in SuS.

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